Update on Merging AIRS, CrIS and IASI Level 2 Temperature and Water Vapor

Eric Fetzer, Peter Kalmus, Hai Nguyen, Yuliya Marchetti, Evan Manning, Amy Braverman, Evan Fishbein, Bjorn Lambrigtsen, Sun Wong, Joao Teixeira and Thomas Pagano

Jet Propulsion Laboratory, California Institute of Technology

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The Goal

A seamless temperature and water vapor (and trace gas) product

- Independent of source instrument(s).
- Resolution, information content, etc. represented numerically.
- Consistent quality flagging
 - Or none at all.

For users: A web search on "NASA Sounder Temperature and Water Vapor" will point to appropriate products

NOTE: The NRC has argued for more than one product.

Overview of Instruments

- AIRS / AMSU / HSB
 - Launched 2002.
 - Standard AIRS retrievals since 2003.
- Two IASI / AMSU / MHS
 - Neural net-based retrievals from EUMETSAT.
 - Bjorn Lambrigtsen and Amy Braverman (JPL) are working with Thomas August (EUMETSAT) on microwave sounder and AIRS-IASI coordination.
- One (almost two) CrIMSS
 - SNPP and JPSS-1 (this calendar quarter).

Sounder Retrieval Data Sets Approximately ten to be combined?

Mix and match any and all:

- NUCAPS (NOAA; Unique Combined Atmospheric Processing System)
 - SNPP only.
- CLIMCAPS (STC)
 - CrIMSS
 - AIRS/AMSU possible.
- CHARTS (GSFC STC)
 - The AIRS standard retrieval.
 - CrIMSS also.
- MW-only NUCAPS
- MW-only (JPL Lambrigtsen)
 - All microwave sounders since 2000.
- IDPS
 - SNPP CrIMSS algorithm delivered at launch (?).
- AER
 - CrIMSS on SNPP.
- Irion (JPL)
 - AIRS currently.
 - Proposed for SNPP.
- EUMETSAT
 - Both IASA instrument neural net.
- UW, LARC, etc....

Common Formats and Consistent Products from Sounders Part 2: Science Issues

Eric Fetzer and Evan Manning
Jet Propulsion Laboratory, California Institute of Technology

NASA Sounder Science Team Meeting
Greenbelt, MD
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From last year

Update: Fall 2016 Sounder STM

Some Questions

- How do we interpret quality flags from non-identical instruments in different orbits?
 - Answer: Still an open question.
- How do we convey vertical resolution from those different instruments?
 - Answer: Averaging kernels.
- What about uncertainties (aside from empirical estimates of relative differences)?
 - Answer: No systematic, consistent approach.
- How do clouds affect the sampling of different IR instruments?
 - Answer: Still TBD given different algorithm approaches.
- What about fundamentally different retrieval algorithms?
 - Answer: Still TBD.

More Than Plans: We are creating a data set

A Merged Temperature and Water Vapor Record from Modern Sounders

A proposal to the Satellite and Calibration Interconsistency Studies NASA Research Announcement (NNH15ZDA001N-SCIS)

Eric Fetzer, PI; Van Dang, Co-I; Sun Wong, Co-I; Evan Fishbein, Collaborator; Steven Friedman, Collaborator; Bjorn Lambrigtsen, Collaborator; Brian Kahn, Collaborator; Baijun Tian, Collaborator' Qing Yue, Collaborator

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Relevant instruments providing temperature and water vapor retrievals (from our proposal)

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Infrared Instrument	AIRS	IASI	CrIS	IASI
mstrument				
Microwave	AMSU-A, HSB	AMSU-A, MHS	ATMS	AMSU-A, MHS
Instruments				
Agency	NASA	EUMETSAT	NOAA	EUMETSAT
Satellite	Aqua	MetOp-A	S-NPP	MetOp-B
Start Date	31 Aug 2002	19 Oct 2006	28 Oct 2011	17 Sep 2012
Equator crossing	1:30 PM	9:30 PM	1:30 PM	9:30 PM
time				
Orbit Period	98 minutes	101 minutes	101 min	101 minutes
Orbit altitude	700 km	817 km	817 km	817 km

Fall 2016 Sounder STM

Ranking of Data Merger Challenges Hard to Easy

- 1. Providing consistent Level 2 (and 3) information content from different instrument suites on separate platforms.
 - Not something our community has done in the past.
 - Much has been done with TOVS and microwave instruments.
- 2. Organizing that information in a way that is useful to ALL users.
 - Many lessons learned already, work is underway.
- 3. Implementing a common data format.
 - Evan Manning is organizing this for hundreds of variables.

Data fusion overview

Slides provided by Peter Kalmus, JPL

Benefits of fusion:

- Data collection is often incomplete, sparse, and yields spatially incompatible results. Our goal is to infer the true process from all available data sources.
- Data fusion can capitalize on complementary strengths to minimize prediction errors.

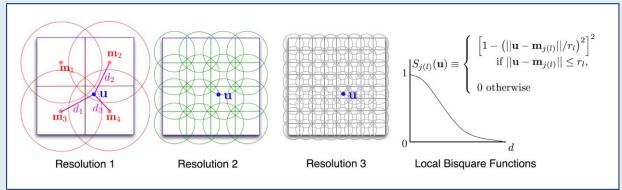


Challenges for data fusion in remote sensing applications:

- Massive size.
- Footprint misalignment.
- Instrument biases.
- Different measurement error characteristics.

Spatial Statistical Data Fusion

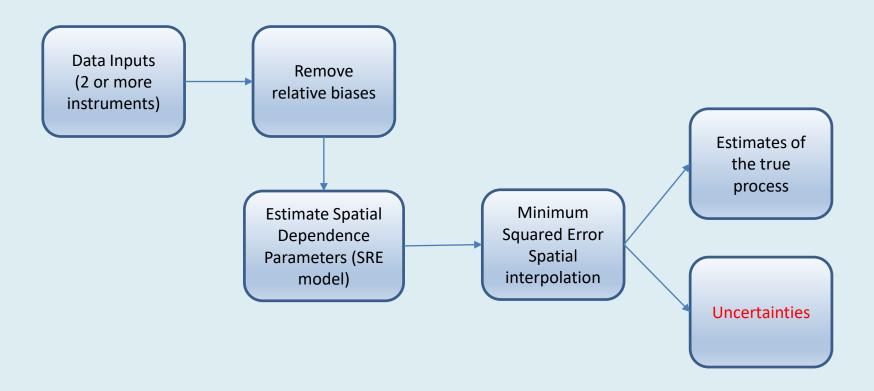
- Our data fusion method addresses the challenges above using a Spatial Random Effects model (SRE; Cressie and Johanneson, 2008)
- It models the spatial dependence using a dimension reduction technique, allowing us to apply spatial interpolation to massive datasets.



Example of basis function in SRE model

• The methodology accounts for spatial dependence, inter-instrument dependence, and different measurement error characteristics (bias and uncertainty)

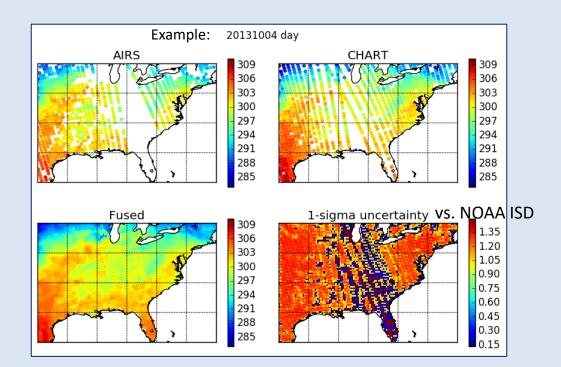
Algorithm flowchart



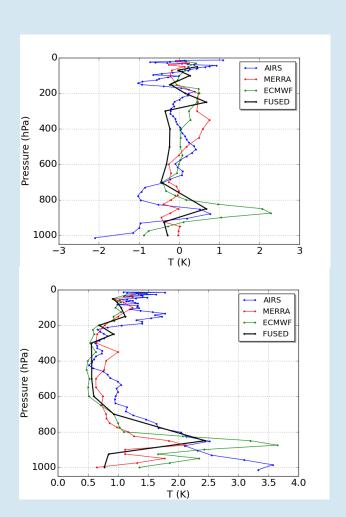
Data Fusion of AIRS + CrIS Near-surface Temperatures

Fusion done on a daily basis, separately for day & night, over Eastern U.S.

- AIRS v6 support product + CrIMSS CHART support product.
- Uses NOAA ISD (Integrated Surface DB) to estimate input errors.
- Could extend to other instruments (e.g. IASI) and to global domain.
- For climate model evaluation, applications (heat waves, drought...).

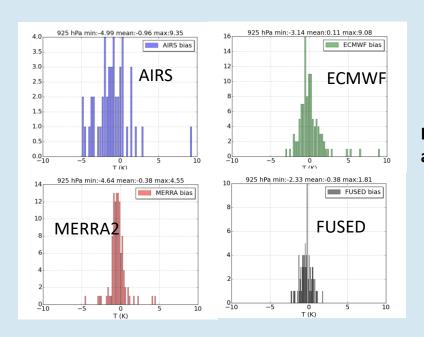


Data Fusion of Temperature Profiles proof of concept



Fusion of 3 data sets: AIRS+ECMWF+MERRA2

- Levels fused separately, T, q fused separately (for now).
- July 2013, 10N—40N, 130W—160W.
- Errors estimated from MAGIC radiosondes
- Next step: fusion in subtropical oceans.
 - Estimate errors using MAGIC + machine learning.



Bias histograms at 925 hPa

Data Fusion References

- Cressie, Noel, and Gardar Johannesson. "Fixed rank kriging for very large spatial data sets." Journal of the Royal Statistical Society: Series B (Statistical Methodology) 70.1 (2008): 209-226.
- Nguyen, Hai, Noel Cressie, and Amy Braverman. "Spatial statistical data fusion for remote sensing applications." *Journal of the American Statistical Association* 107.499 (2012): 1004-1018.
- Nguyen, Hai, et al. "Spatio-temporal data fusion for very large remote sensing datasets." Technometrics 56.2 (2014): 174-185.
- Nguyen, Hai, Noel Cressie, and Amy Braverman. "Multivariate spatial data fusion for very large remote sensing datasets." Remote Sensing 9.2 (2017): 142.
- Kalmus, Peter, Hai Nguyen, Amy Braverman, Yuliya Marchetti, and Evan Fishbein., "Spatial statistical data fusion of AIRS and CrIS near surface temperature." In preparation.

Summary and Conclusions

Summary

- More than a decade of overlapping hyperspectral IR sounder coverage.
 - Even longer for microwave instruments.
- Still many issues to be resolved to merge L2 and L3.
- A long history of data fusion in the statistical literature.
 - E. g. kriging.
- Making progress on how to convey common information.

Conclusions

- Consistent retrieval algorithm(s) will simplify data fusion.
 - This will require L2 developed to characterize long-term behavior.